

Image processing with scikit-image

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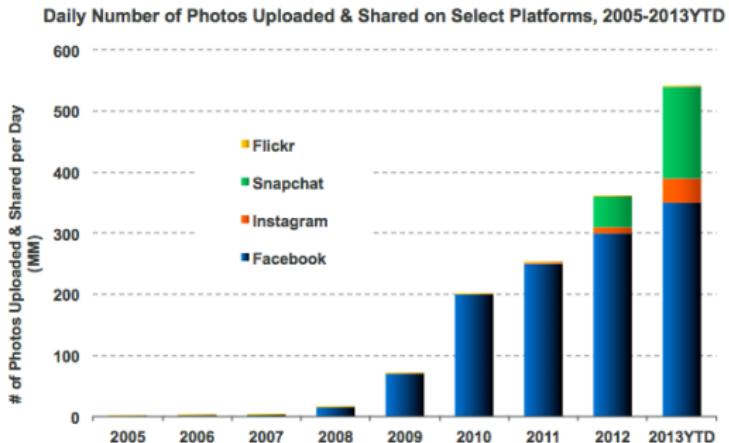
Paris-Saclay Center for Data Science



scikit-image
image processing in python

The world is getting more and more visual

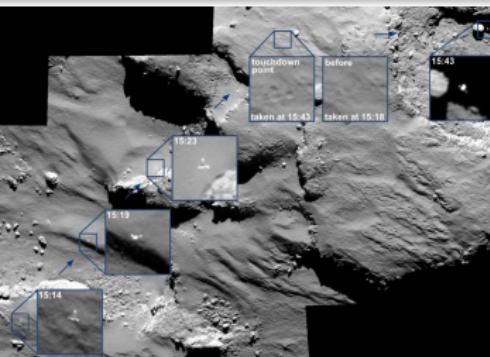
**Photos = 500MM+ Uploaded & Shared Per Day,
Growth Accelerating, on Trend to Rise 2x Y/Y...**



KPCB

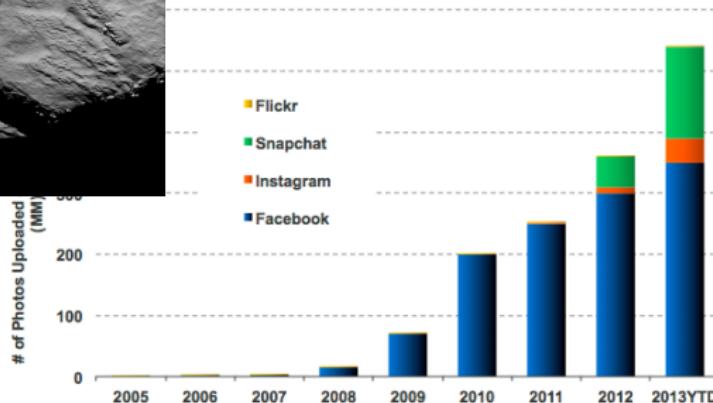
Source: KPCB estimates based on publicly disclosed company data. 14

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MM+ Uploaded & Shared Per Day,
Accelerating, on Trend to Rise 2x Y/Y...

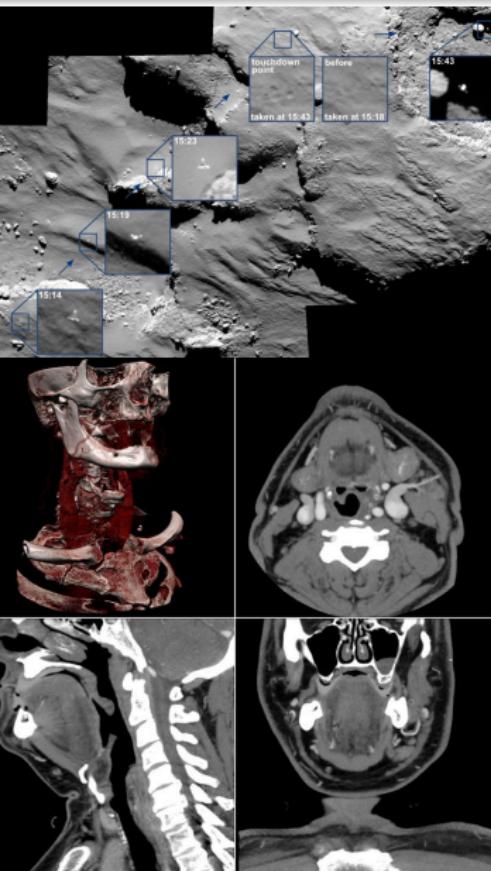
of Photos Uploaded & Shared on Select Platforms, 2005-2013YTD



KPCB

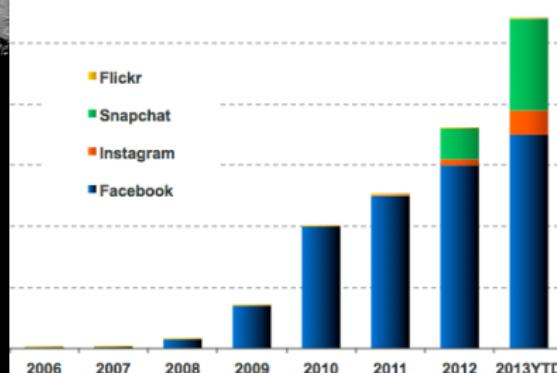
Source: KPCB estimates based on publicly disclosed company data. 14

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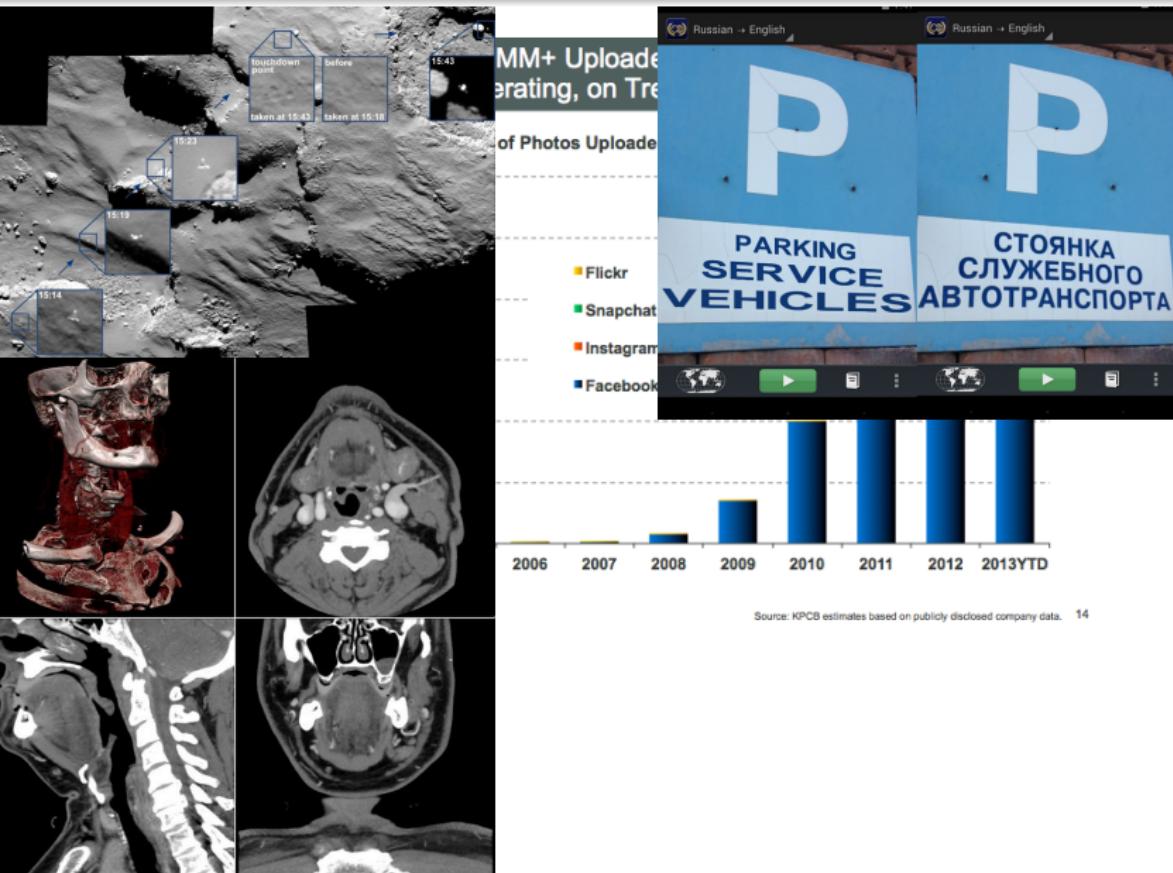
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of Photos Uploaded & Shared on Select Platforms, 2005-2013YTD

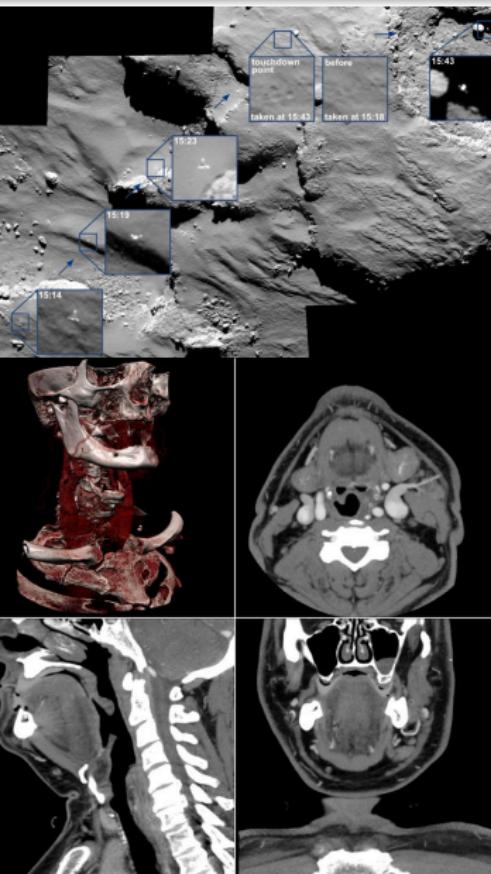


Source: KPCB estimates based on publicly disclosed company data. 14

The world is getting more and more visual



The world is getting more and more visual



The world is getting more and more visual



Image processing

- Manipulating images in order to retrieve new images or image characteristics (features, measurements, ...)
- Often combined with machine learning

scikit-image

<http://scikit-image.org/>

A module of the Scientific Python stack

- Language: Python
 - Core modules: NumPy, SciPy, matplotlib
 - Application modules: scikit-learn, scikit-image, pandas, ...

A general-purpose image processing library

- open-source (BSD)
- not an application (ImageJ)
- less specialized than other libraries (e.g. OpenCV for computer vision)

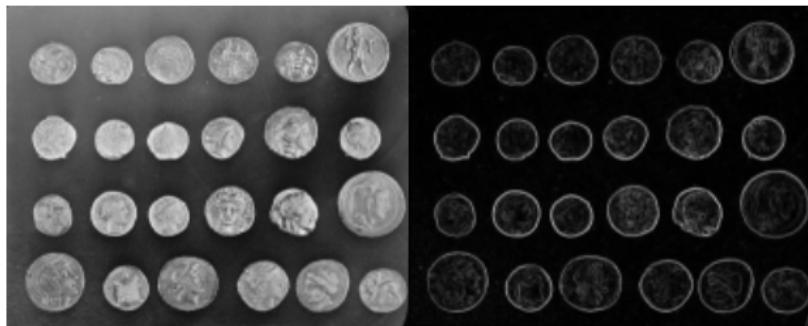


scikit-image
image processing in python

1 Principle

1 First steps

```
from skimage import data, io, filter  
  
image = data.coins() # or any NumPy array!  
edges = filter.sobel(image)  
io.imshow(edges)  
io.show()  
x
```



My environment: IPython interpreter + text editor
Ipython notebook nice for demos/trial and error

1 Manipulating images as numpy arrays

- numpy arrays as arguments and outputs

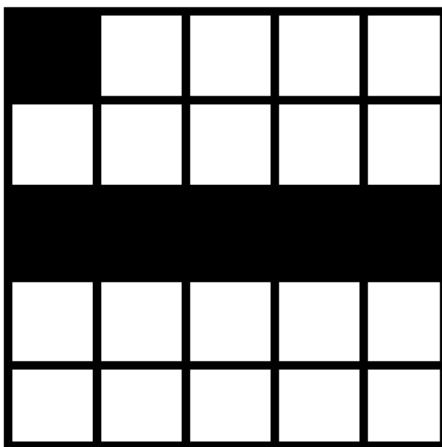
```
>>> from skimage import io, filter  
>>> camera_array = io.imread('camera_image.png')  
>>> type(camera_array)  
<type 'numpy.ndarray'>  
>>> camera_array.dtype  
dtype('uint8')  
>>> filtered_array = filter.gaussian_filter(  
    camera_array, sigma=5)  
>>> type(filtered_array)  
<type 'numpy.ndarray'>  
>>> filtered_array.dtype  
dtype('float64')
```



1 Manipulating images as numerical (numpy) arrays

- Pixels are arrays elements

```
import numpy as np  
image = np.ones((5, 5))  
image[0, 0] = 0  
image[2, :] = 0
```



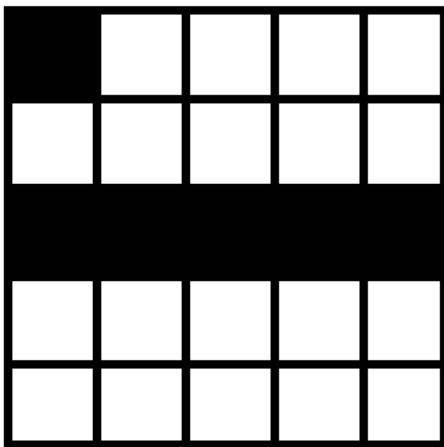
(use matplotlib for visualization: `matplotlib.pyplot.imshow`)

1 Manipulating images as numerical (numpy) arrays

- Pixels are arrays elements

```
import numpy as np  
image = np.ones((5, 5))  
image[0, 0] = 0  
image[2, :] = 0
```

X



(use matplotlib for visualization: `matplotlib.pyplot.imshow`)

1 Some magics inside

Don't let yourself be tricked by integer / float conversion!

```
>>> from skimage import data, filter
>>> camera_array = data.camera()
>>> camera_array.dtype
dtype('uint8')
>>> filtered_array = filter.gaussian_filter(
    camera_array, sigma=5)
>>> filtered_array.dtype
dtype('float64')
>>> camera_array.min(), camera_array.max()
(0, 255)
>>> filtered_array.min(), filtered_array.max()
(0.031287384322526979, 0.8560994897846772)
```



1 An API relying mostly on functions

```
skimage.filter.gaussian_filter(image, sigma, output=None, mode='nearest', cval=0, multichannel=None)
```

Multi-dimensional Gaussian filter

Parameters

image : array-like

image (grayscale or color) to **filter**.

sigma : scalar or sequence of scalars

 standard deviation for Gaussian kernel. The
 standard

 deviations of the Gaussian **filter** are given for
 each axis as a

 sequence, or as a single number, in which case it
 is equal for

 all axes.

output : array, optional

 The “‘output’‘ parameter passes an array in which
 to store the
 filter output.

1 Images and dimensions

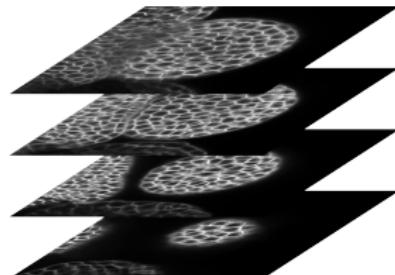


```
>>> data.camera().shape  
(512, 512)
```

- Most functions suitable for 2-D gray- or color-scale images
- Some functions work with 3D images as well
- Check out `scipy.ndimage` for n-d functionnality.



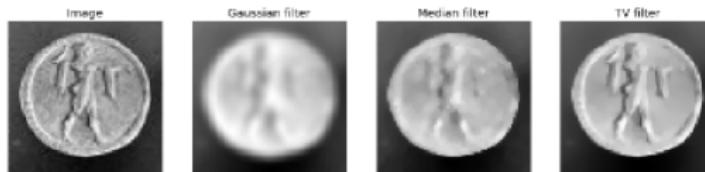
```
>>> coffee.shape  
(400, 600, 3)  
>>> red_channel =  
coffee[..., 0]
```



(d_0, d_1, d_2)

2 Some features

2 Filtering: transforming image data

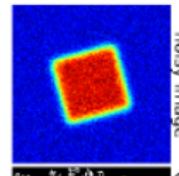


Impact

Gaussian filter

Median filter

TX filter



noisy images

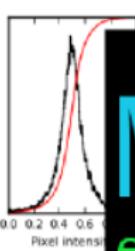
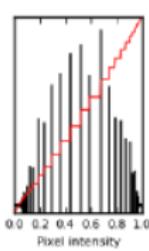
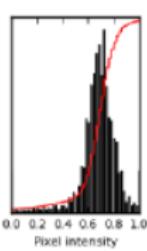
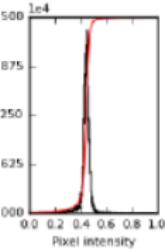
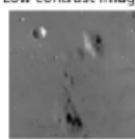
Canny filter

Canny filter, $\sigma = 3$



Contrast stretching

Histogram equalization



denoising sobel
equalize wiener
Median
Gaussian canny
enhance _ contrast
total variation

`skimage.filter`, `skimage.exposure`, `skimage.restoration`

2 From very simple/classical algorithms

noisy



Gaussian filter



Median filter



```
from skimage import data, filter, color
from skimage.morphology import disk

I = data.lena()
I = color.rgb2grey(I)
I = I[230:290, 220:320]

noisy = I + 0.4 * I.std() * np.random.random(I.shape)

gaussian_denoised = filter.gaussian_filter(noisy,
    sigma=2)
median_denoised = filter.rank.median(noisy, disk(3))
```

2 To more advanced/recent algorithms

noisy



TV denoising



(more) TV denoising

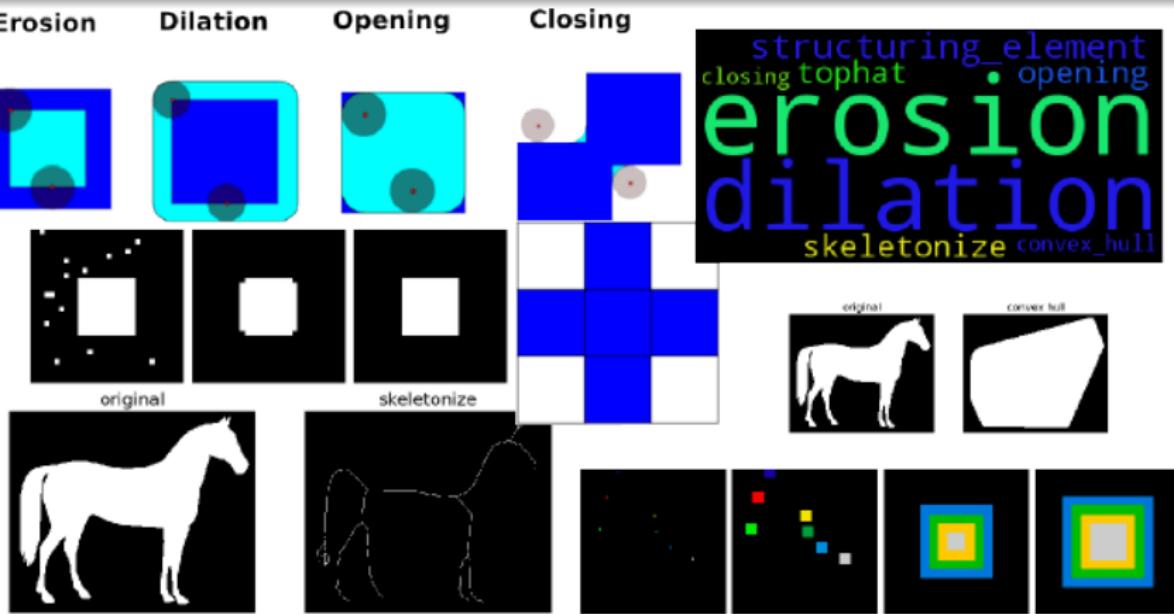


```
from skimage.filter import tv_denoise
from skimage import data

I = data.lena()
I = I[230:290, 220:320]
noisy = I + 0.4*I.std()*np.random.random(I.shape)

tv_denoised = tv_denoise(noisy, weight=10)
```

2 Mathematical morphology



`skimage.morphology`

2 Extracting features

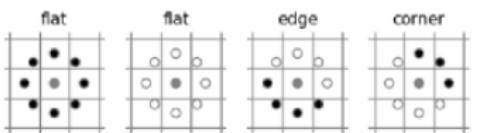
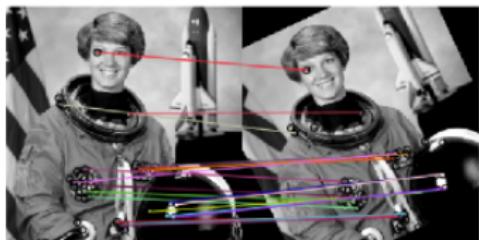
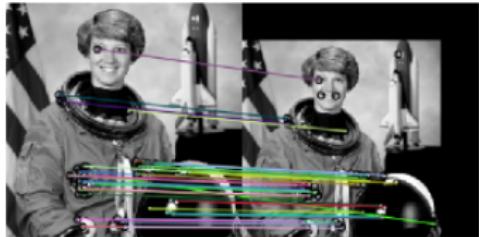
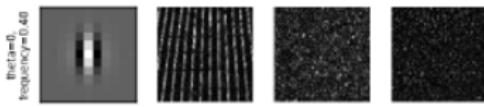
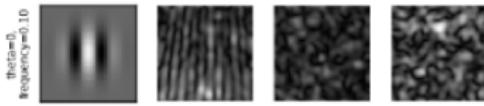


Image responses for Gabor filter kernels



corners
local_maxima
Gabor
canny
Harris
hog
hough
cooccurrence

`skimage.feature`, `skimage.filter`

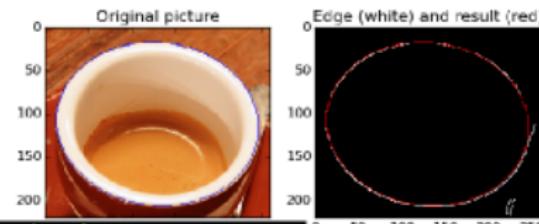
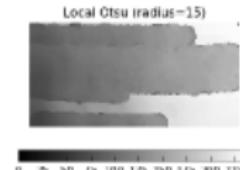
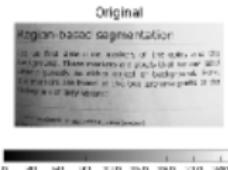
2 Geometrical transformations

`skimage.transform`

scale, zoom, rotate, swirl, warp, ...



2 Segmentation: labelling regions



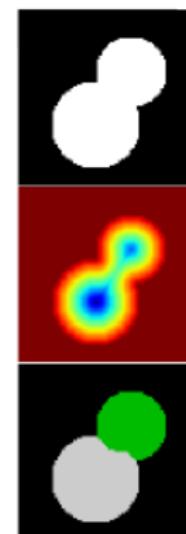
Original >= Local Otsu

Region-based segmentation



watershed
superpixel
thresholding
otsu
randomwalker

Felzenszwalbs's method



`skimage.segmentation`

2 Feature extraction followed by classification

Combining scikit-image and scikit-learn

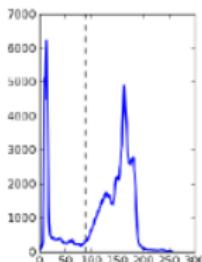
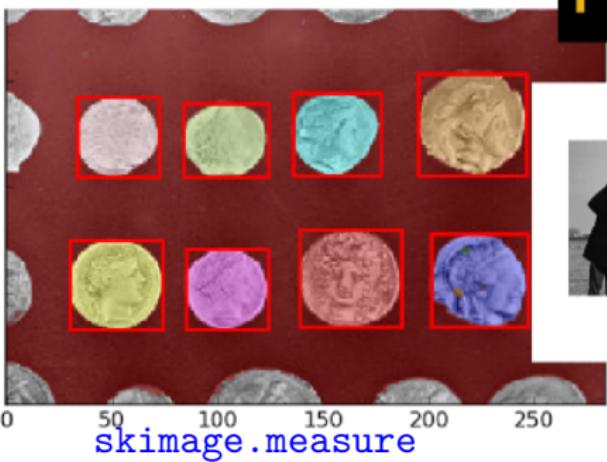
- Extract features (`skimage.feature`)
 - Pixels intensity values (R, G, B)
 - Local gradients
 - More advanced descriptors: HOGs, Gabor, ...
- Train classifier with known regions
 - here, random forest classifier
- Classify pixels



2 Measures on images

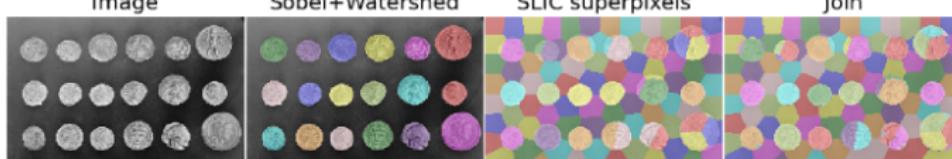
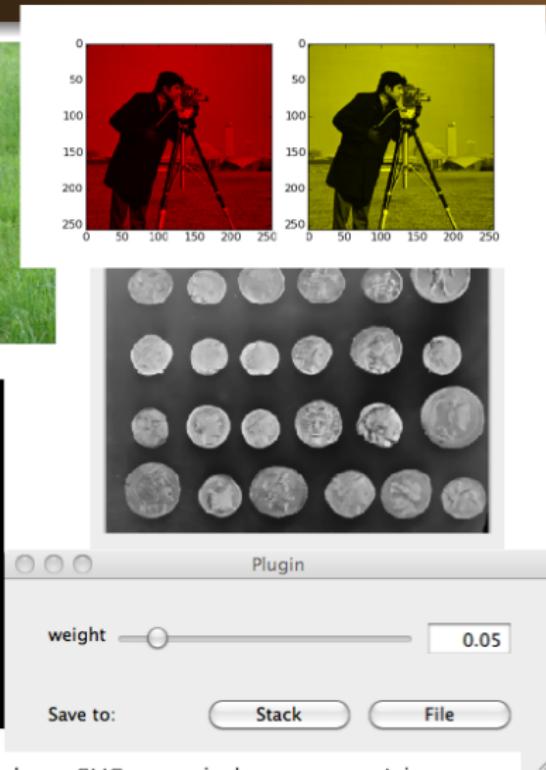
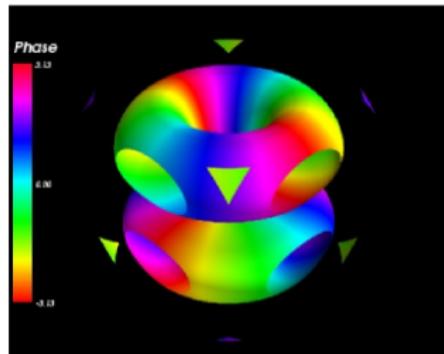
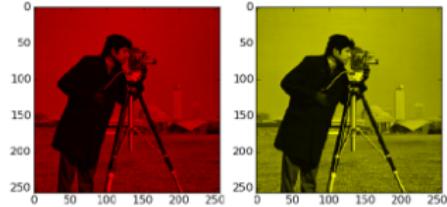


size
label
measure
histogram
regionprops



2 Visualizing images and more

matplotlib, mayavi



2 Development of scikit-image

- Mature algorithms
- Only Python + Cython code for easier maintainability
- Hosted on GitHub
- Thorough code review by others: readability, PEP8, efficiency, ...
- 1-2 releases per year
- Core team of 5 persons (+ GSoc students)



2 Getting started: installing scikit-image

<http://scikit-image.org/docs/dev/install.html>

- Packaged on Ubuntu/Debian
- Shipped with all major Scientific Python distributions:
Enthought Canopy, Anaconda, Python(x,y)

2 Getting started: finding documentation

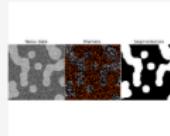


scikit-image
image processing in python™

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Image processing in Python

scikit-image is a collection of algorithms for image processing. It is available **free of charge** and **free of restriction**. We pride ourselves on high-quality, peer-reviewed code, written by an active **community of volunteers**.

[Download](#)

Stable

0.10.1 - June 2014

[Download](#)

Development

pre-0.11

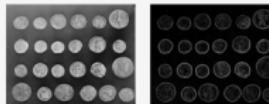
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8+ 307 ⚡ Star 522

Getting Started

Filtering an image with scikit-image is easy! For more examples, please visit our [gallery](#).

```
from skimage import data, io, filter
image = data.coins() # or any NumPy array!
edges = filter.sobel(image)
io.imshow(edges)
io.show()
```



If you find this project useful, please cite:

[BIBTeX](#)

Stéfan van der Walt, Johannes L. Schönberger, Juan Nunez-Iglesias, François Boulogne, Joshua D. Warner, Neil Yager, Emmanuelle Gouillart, Tony Yu and the scikit-image contributors. scikit-image: Image processing in Python. PeerJ 2:e453 (2014) <http://dx.doi.org/10.7717/peerj.453>

Links

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Announcements

2 Getting started: finding documentation



scikit-image
image processing in python™

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API Reference

- **skimage**
 - Subpackages
 - Utility Functions
 - `dtype_limits`
 - `img_as_bool`
 - `img_as_float`
 - `img_as_int`
 - `img_as_ubyte`
 - `img_as_uint`
 - `test`
- **Module: color**
 - `combine_stains`
 - `convert_colorspace`
 - `deltaE_cie76`
 - `deltaE_ciede2000`
 - `deltaE_ciede94`
 - `deltaE_cmc`
 - `gray2rgb`
 - `guess_spatial_dimensions`
 - `hed2rgb`
 - `hsv2rgb`
 - `lab2lch`
 - `lab2rgb`
 - `lab2xyz`
 - `label2rgb`
 - `lch2lab`
 - `luv2rgb`
 - `luv2xyz`
 - `rgb2gray`
 - `rgb2grey`
 - `rgb2hed`
 - `rgb2hsv`
 - `rgb2lab`
 - `rgb2luv`

threshold_otsu

`skimage.filter.threshold_otsu(image, nbins=256)`

Return threshold value based on Otsu's method.

Parameters:

`image` : array

Input image.

`nbins` : int, optional

Number of bins used to calculate histogram. This value is ignored for integer arrays.

Returns:

`threshold` : float

Upper threshold value. All pixels intensities that less or equal of this value assumed as foreground.

References

 [Wikipedia](http://en.wikipedia.org/wiki/Otsu%27s_Method), http://en.wikipedia.org/wiki/Otsu%27s_Method

Examples

```
>>> from skimage.data import camera
>>> image = camera()
>>> thresh = threshold_otsu(image)
>>> binary = image <= thresh
```

threshold_yen

`skimage.filter.threshold_yen(image, nbins=256)`

Return threshold value based on Yen's method.

Parameters:

`image` : array

Input image.

`nbins` : int, optional

2 Gallery of examples



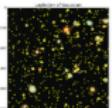
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Search documentation ...

General examples

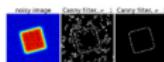
General-purpose and introductory examples for the scikit-image.



Blob Detection



BRIEF binary descriptor



Canny edge detector



CENSURE feature
detector



Circular and Elliptical
Hough Transforms



Contour finding



Convex Hull



Corner detection



Dance DAISY feature

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Versions

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[skimage 0.9.x](#)

[skimage 0.8.0](#)

[skimage 0.7.0](#)

[skimage 0.6](#)

[skimage 0.5](#)

[skimage 0.4](#)

[skimage 0.3](#)

2 Getting started: finding documentation

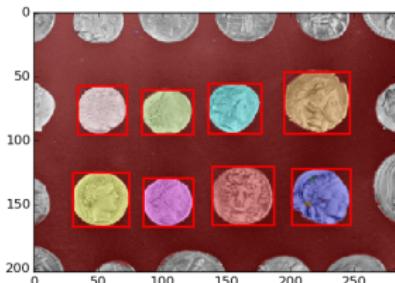


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Label image regions

This example shows how to segment an image with image labelling. The following steps are applied:

1. Thresholding with automatic Otsu method
2. Close small holes with binary closing
3. Remove artifacts touching image border
4. Measure image regions to filter small objects



```
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.patches as mpatches

from skimage import data
from skimage.filter import threshold_otsu
from skimage.segmentation import clear_border
from skimage.morphology import label, closing, square
from skimage.measure import regionprops
from skimage.color import label2rgb

image = data.coins()[50:-50, 50:-50]

# apply threshold
thresh = threshold_otsu(image)
bw = closing(image > thresh, square(3))

# remove artifacts connected to image border
cleared = bw.copy()
clear_border(cleared)

# label image regions
label_image = label(cleared)
borders = np.logical_xor(bw, cleared)
label_image[borders] = -1
image_label_overlay = label2rgb(label_image, image=image)

fig, ax = plt.subplots(ncols=1, nrows=1, figsize=(6, 6))
ax.imshow(image_label_overlay)

for region in regionprops(label_image):

    # skip small images
    if region.area < 100:
        continue

    # draw rectangle around segmented coins
    minc, minr, maxc, maxr = region.bbox
    rect = mpatches.Rectangle(minc, minr, maxc - minc, maxr - minr,
                             fill=False, edgecolor='red', linewidth=2)
    ax.add_patch(rect)

plt.show()
```

2 Conclusions

scikit-image

- An image processing Python module relying on NumPy arrays
- Trade-off between performance and usability
- More and more features
- Try it out!
 - <http://scikit-image.org/>
 - <https://www.youtube.com/watch?v=SE7h0IWD93Y>
(and others)
 - <http://scipy-lectures.github.io/packages/scikit-image/>